Fleet Mixture and Arrival Rate Estimation at Memphis International Airport

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Presentation Goals

- Demonstrate applicability of DROMS database at Memphis International Airport (MEM)
- DROMS = Dynamic RunwayOccupancy Measurement System
- Estimate parameters for airport arrival capacity modeling at MEM





Objectives

- From January 2003 data (14,252 arrivals) at MEM, estimate:
 - IMC frequency
 - Aircraft weight class percentages
 - Dominant airline fleet compositions
 - Frequencies of aircraft weight class pairs
 - Arrival rates for lead/trail aircraft weight class pairs by regression
 - Effects on capacity modeling





DROMS Database

Collection at MEM and DTW ongoing:

| Airport | Operations | Days |
|---------|------------|------|
| MEM | 224,283 | 215 |
| DTW | 518,455 | 435 |

- Collection at STL planned in mid-2004
- Planned distribution by Sensis Corporation,
 NASA, and Volpe Research Center





DROMS Data Sources

- Multilateration position data
 - Cooperative (secondary) surveillance
 - Surface and terminal coverage
 - Surface coverage (MEM, DTW)
 - Terminal coverage (MEM)
- Aviation Situation Display to Industry (ASDI)
- Runway Visual Range (RVR)
- Aviation Routine Meteorological Report (METAR)
- Aviation System Performance Metrics (ASPM)
- Aircraft physical properties
- Aircraft registration data





VMC vs. IMC at MEM

- Analysis of ASPM and flight data:
 - Quarter-hour ASPM data for Jan 2003
 - 73% of meteorological reports are VMC
 - 71% of arrivals landed under VMC
- Conclusion:
 - Arrivals can be segregated by meteorological condition without adding bias to data subsets





MEM Weight Class Mixture

- FAA Wake Vortex Weight Classes
 - ICAO classification also available
- 14,252 arrivals in January 2003:

large: 40.9%

heavy: 30.8%

small: 21.8%

B757: 1.4%

unknown: 5.1%





Airline Fleet Mixture at MEM

Dominant airlines:

75%

Northwest Airlines and

Express Airlines:

97% large

Mesaba Airlines:

100% small

Federal Express:

99.6% heavy

Other airlines and GAs: 25%

52% small, 48% large





Observed Pair Percentages at MEM

| 12,844 pairs | | Lead Aircraft | | | | |
|-------------------|-------|---------------|-------|-------|------|-------|
| | | small | large | heavy | B757 | sum |
| Trail Aircraft | small | 6.4 | 12.9 | 3.5 | 0.3 | 23.1 |
| | large | 13.0 | 25.2 | 4.5 | 0.8 | 43.5 |
| | heavy | 3.4 | 4.7 | 23.8 | 0.2 | 32.0 |
| | B757 | 0.4 | 0.8 | 0.2 | 0.1 | 1.5 |
| | sum | 23.2 | 43.5 | 31.9 | 1.4 | 100.0 |





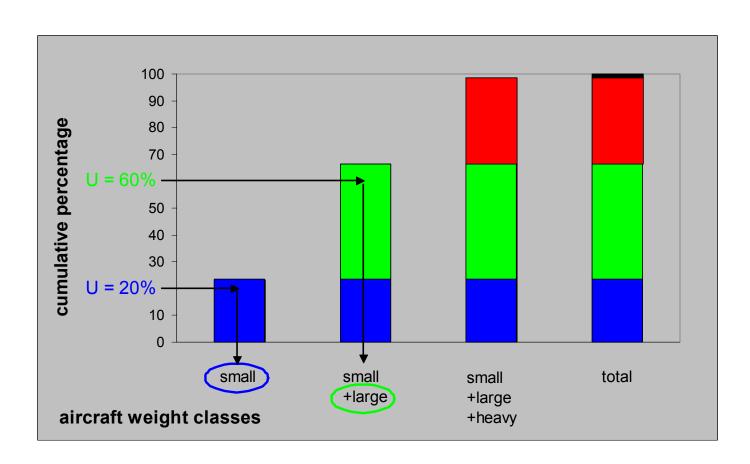
Independent Pair Simulation

- Simulate expected lead/trail weight class percentages:
 - Lead and trail aircraft weights classes occur independently
 - Lead and trail aircraft weight class pairs are selected from observed weight class percentages (% small aircraft)
 - Averaged over 2.5 x 10⁶ Monte Carlo realizations





Weight Class Selection







Independent Pair Percentages (Simulated) at MEM

| | | Lead Aircraft | | | | |
|-------------------|-------|---------------|-------|-------|------|-------|
| | | small | large | heavy | B757 | sum |
| Trail Aircraft | small | 5.4 | 10.1 | 7.4 | 0.3 | 23.1 |
| | large | 10.0 | 18.9 | 13.9 | 0.6 | 43.4 |
| | heavy | 7.4 | 13.9 | 10.3 | 0.5 | 32.0 |
| | B757 | 0.3 | 0.6 | 0.5 | <0.1 | 1.4 |
| | sum | 23.1 | 43.4 | 32.0 | 1.4 | 100.0 |





Conditional Pair Simulation

Simulate expected lead/trail weight class percentages:

- Trail aircraft weight class is conditioned on lead aircraft weight class
- Markov chain model implemented using observed cumulative weight class pairs





Conditional Pair Percentages (Simulated) at MEM

| | | Lead Aircraft | | | | |
|-------------------|-------|---------------|-------|-------|------|-------|
| | | small | large | heavy | B757 | sum |
| Trail Aircraft | small | 6.4 | 12.9 | 3.5 | 0.4 | 23.1 |
| | large | 13.0 | 25.1 | 4.5 | 0.8 | 43.4 |
| | heavy | 3.3 | 4.7 | 23.8 | 0.2 | 32.0 |
| | B757 | 0.4 | 0.8 | 0.2 | 0.1 | 1.5 |
| | sum | 23.1 | 43.4 | 31.1 | 1.5 | 100.0 |





Independent vs. Conditional Pairs

• Independent vs. observed: $\sigma = 5.28$

• Conditional vs. observed: $\sigma = 0.03$

- Simulation of single runway capacity (after Lang et al., 2003)
 - Independent: 25.8 arrivals per hour
 - Conditional: 26.5 arrivals per hour





Arrival Rate Regression Models

- Measure quarter-hourly arrival rate
- Measure inter-arrival distance and time spacings between aircraft pairs (grouped by runway)
- Separate by VMC and IMC
- Regress arrival rate on distance and time spacing for lead/trail weight class pairs
- $log(arr rate) = b_0 + b_1 log(spacing)$





Regression Results:

- Arrival Rate vs. Distance Spacing
 - VMC: all but B757 pairs, heavy/small
 - IMC: 5 pair combinations
- Arrival Rate vs. Time Spacing
 - VMC: all but B757 pairs
 - IMC: 6 pair combinations
- Estimate arrival rates from inter-arrival distance or time by weight class pairs





Application of Regression Models

- FAA distance spacing table
- VMC regression equations to predict arrival rate by weight class pair
- Weight class pair percentages:
 - Independent simulations
 - Conditional simulations
- Arrival Rate Estimates:

Independent: 36.3 per hour

Conditional: 37.0 per hour





Summary, 1

14,252 Arrivals to MEM, Jan. 2003

Estimated VMC vs. IMC frequency

Fleet mixtures of dominant airlines





Summary, 2

- Weight class percentages:
 - Observed
 - Independent pairing model
 - Conditional pairing model
- Regression models to estimate arrival rates for weight class pair:
 - inter-arrival distance spacing data
 - inter-arrival time spacing data





Future Research

- Extend analysis to include more time data
- Extend analysis to include other airports (DTW, STL)
- Restrict analysis to time periods with high arrival rates
- Identify and estimate statistics of inter-arrival distance and time spacing distributions
- Further evaluate impact on capacity modeling with distance and time spacings based on statistical distributions



